

OPTIMIZING TOMOPY

Performance analysis of grid reconstruction

Benchmarking: make_data.py

```
import numpy as np, tomopy

obj = tomopy.shepp3d(size=512)

ang = tomopy.angles(750) # Generate uniformly spaced tilt angles.

sim = tomopy.project(obj, ang) # Calculate projections.

np.save('projection.npy', sim)

np.save('angles.npy', ang)
```

Checking dimensions and type of the projection data:

```
In [1]: import numpy as np
In [2]: sim = np.load('projection.npy')
In [3]: sim.shape
Out[3]: (750, 512, 728)
In [4]: sim.dtype
Out[4]: dtype('float32')
```



Tomopy out of the box

```
conda create --name tomopy_nomkl \
   -c dgursoy \
   nomkl tomopy pyfftw fftw numpy scipy numexpr pywavelets \
   scikit-image ipython ipython-notebook astropy \
   python=3.5

conda create --name tomopy \
   -c dgursoy \
   tomopy pyfftw fftw numpy scipy numexpr pywavelets \
   scikit-image ipython ipython-notebook astropy \
   python=3.5
```



Reconstruction script

```
import numpy as np, tomopy, time as t
    def timeit(func, named_args, kwargs):
        t0 = t.time()
        r = func(*named args, **kwargs)
        t1 = t.time()
 6
        return (t1 - t0, r)
 8
    sim = np.load('projection.npy')
 9
10
    ang = np.load('angles.npy')
11
12
    # Reconstruct object:
    recon time, rec = timeit(tomopy.recon, (sim, ang), dict(algorithm='gridrec'))
13
14
    print("Reconstruction time: {0:.3f}".format(recon_time))
15
```

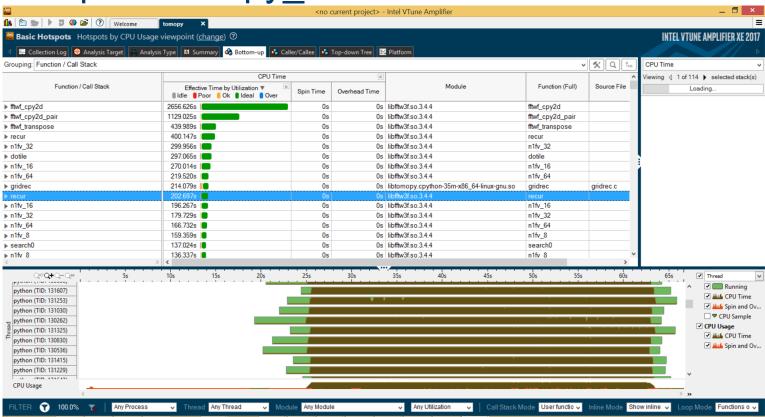


Performance times

(knl)\$ numactl - python reco				
IZNII	n a nald	256	47.606	
KNL	nomkl	256	47.696	
KNL	mkl	256	98.56	must set OMP_NUM_THREADS=1
KNL	mkl	256	12.965	KMP_AFFINITY=disabled
(hsw)\$ python				
recon_bench.py				
HSW	nomkl	32	4.246	
HSW	mkl	32	11.356	
HSW	mkl_seq	32	3.294	MKL_THREADING_LAYER=SEQUENTIAL

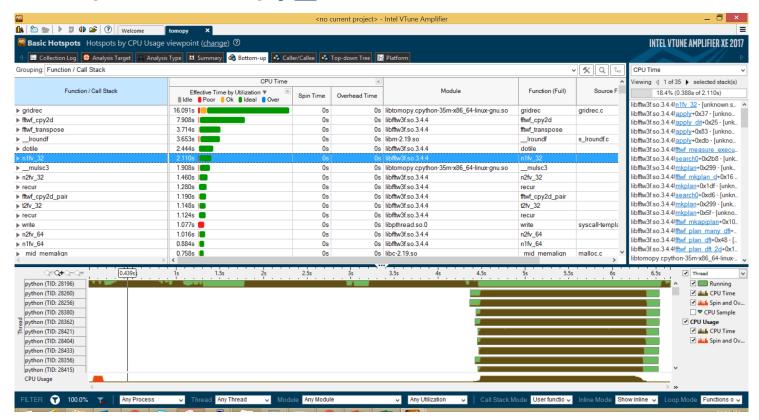


Hotspots tomopy_nomkl on KNL





Hotspots tomopy_nomkl on HSW





Building tomopy toolchain with icc

- Created recipes to build essential components with icc targeting common-avx512 architechture
- Small changes to tomopy itself
 - removed –Im in setup.py, vectorized code in phantom.py
 - changes to gridrec.c to enable vectorization
- Modules are fftw, pyfftw, tomopy, dxchange, dxfile, olefile are built locally
- Modules numpy, scipy, scikit-image are conda-installed from intel channel
- Other modules (pywavelets, etc) taken from dgursoy channel
- netCDF4 and atropy were pip or conda installed



Tomopy recipe: build.sh

```
#!/bin/bash

export CC=icc

export LDSHARED="icc -shared"

$PYTHON setup.py config

C_INCLUDE_PATH="$PREFIX/include" \

LD_LIBRARY_PATH="$PREFIX/lib" \
CFLAGS="-m64 -fomit-frame-pointer -pthread -qopenmp -fPIC -fp-model fast=2 -03 -xCORE-AVX2 -axCOMMON-AVX512 -I$PREFIX/include $CFLAGS" \
LDFLAGS="-L$PREFIX/lib $LDFLAGS" \
$PYTHON setup.py build_ext --inplace

$PYTHON setup.py install --old-and-unmanageable
```

Compile tomopy using icc targeting both HSW and KNL, enabling vectorization.

Recipes are available on cori.

Used vectorization report (–qopt-report=5) to guide optimizations



Tomopy recipe, cont.

```
{% set version = "1.0.1" %}
    {% set buildnumber = 6 %}
     {% set iccver = "16.0.3" %}
                                     [unix or py35]
     package:
       name: tomopy
       version: {{version}}
     build:
       number: {{buildnumber}}
10
11
       features:
12
         - intel
13
14
     source:
15
         git_url: https://github.com/tomopy/tomopy
16
         git rev: master
         patches:
17
18
           - intel changes.patch
19
20
    requirements:
21
         build:
           - python
22
23
          - intelpython
24
           - icc rt
25

    setuptools

26
           - numpy
           - fftw
27
```

Patch represents diff between official github.com/tomopy/tomopy.git and branch **feature/intelem** of its fork github.com/oleksandr-pavlyk/tomopy.git

Gist of optimizations

- Replace Iroundf(x) with (int) roundf(x)
- Replace ceil(x) with ceilf(x), etc.
- Replace fabs(x) with fabs(f)
- Apply vectorization pragmas
- Split one double loop to enable vectorization



Changes in gridrec.c

```
for(iu=iul; iu<=iuh; iu++)
{
    rtmp = wtbl[lroundf(fabs(U-iu)*tblspcg)];
    for(iv=ivl, k=0; iv<=ivh; iv++, k++)
    {
        const float convolv = rtmp*work[k];
        H[iu][iv] += convolv*Cdata1;
        H[pdim-iu][pdim-iv] += convolv*Cdata2;
    }
}</pre>
```

```
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```

```
iuh2 = (pdim2 > iuh) ? iuh : pdim2 - 1;
#pragma simd assert
       for(iu=iul; iu <= iuh2; iu++)</pre>
           rtmp = wtbl[(int) roundf(fabsf(U-iu)*tblspcg)];
           for(iv=iv1, k=0; iv<=ivh; iv++, k++)</pre>
               const float convolv = rtmp*work[k];
               H[iu][iv] += convolv*Cdata1;
               H[pdim-iu][pdim-iv] += convolv*Cdata2;
       // assert( iu == pdim2 || iu > iuh );
       for( ; iu <= pdim2 && iu <= iuh; iu++)</pre>
           rtmp = wtbl((int) roundf(fabsf(U-iu)*tblspcg));
           for(iv=iv1, k=0; iv<=ivh; iv++, k++)</pre>
               const float convolv = rtmp*work[k]:
               H[iu][iv] += convolv*Cdata1;
               H[pdim-iu][pdim-iv] += convolv*Cdata2;
#pragma simd assert
       for(: iu<=iuh: iu++)
           rtmp = wtbl[(int) roundf(fabsf(U-iu)*tblspcg)];
           for(iv=iv1, k=0; iv<=ivh; iv++, k++)</pre>
               const float convolv = rtmp*work[k];
               H[iu][iv] += convolv*Cdata1;
               H[pdim-iu][pdim-iv] += convolv*Cdata2;
```

Building tomopy-recipe

```
#!/bin/bash
export CONDA_BLD_PATH=./conda-build

conda build -c intel -c dgursoy --override-channels \
--no-anaconda-upload --python 3.5 --numpy 1.11 tomopy-recipe
```



Using built tomopy

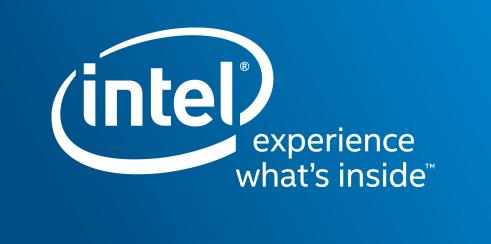
```
#!/bin/bash -x
 2
    export ENV =$1
 4
    conda create --name $ ENV -c intel numpy scipy scikit-image numexpr h5py hdf5 six ipython python=3.5 --yes
    source activate $ ENV
    conda install -c intel -c dgursoy --override-channels pywavelets tifffile edffile spefile --yes
    conda install astropy --yes
    pip install netCDF4
10
11
    # install modules locally built with icc
     pushd conda-build/linux-64
12
     conda install \
13
        dxchange-0.1.2-py35 intel 0.tar.bz2 \
14
        fftw-3.3.6-intel 1.tar.bz2 \
15
16
        pvfftw-0.10.4-pv35 intel 0.tar.bz2 \
        dxfile-0.4.0-py35 intel 0.tar.bz2 \
17
        olefile-0.44.0-py35 intel 1.tar.bz2 \
18
        tomopy-1.0.1-py35_intel_6.tar.bz2
19
20
```



Performance results

python reco	on_bench.py			
HSW	optimized	32	1.343	
KNL	optimized	256	2.492	KMP_AFFINITY=disabled numactl -p 1 python recon_bench.py





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